# What is MLOps?

**MLOPS CONCEPTS** 



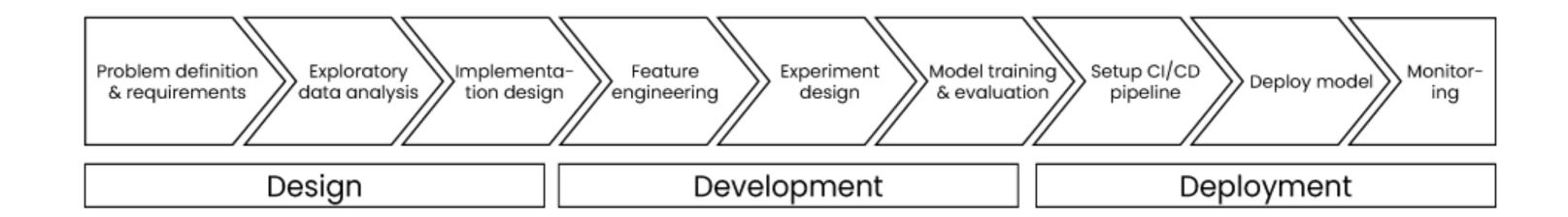
Folkert Stijnman ML Engineer



### Machine Learning Operations

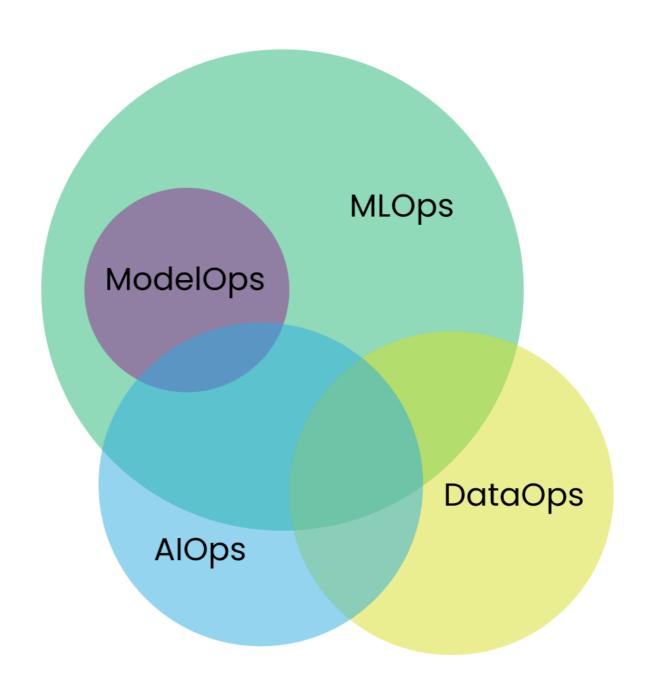
...is the set of practices to design, deploy and maintain machine learning in production continuously, reliably, and efficiently.

- Focus on machine learning 'in production'
- The full machine learning lifecycle

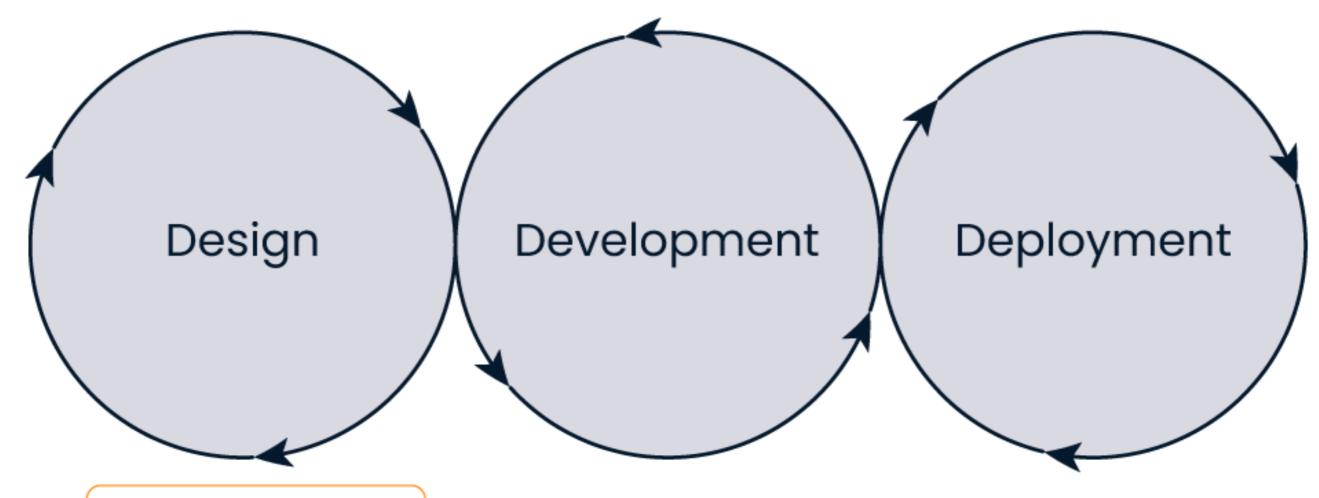


### WhatOps?

- ModelOps: Model Operations
  - Is primarily focused on the machine learning model
- DataOps: Data Operations
  - Focuses on best practices in data quality and analytics
- AlOps: Artificial Intelligence for IT Operations
  - Focuses on using analytics, big data, and machine learning to solve IT issues
    without human assistance or intervention



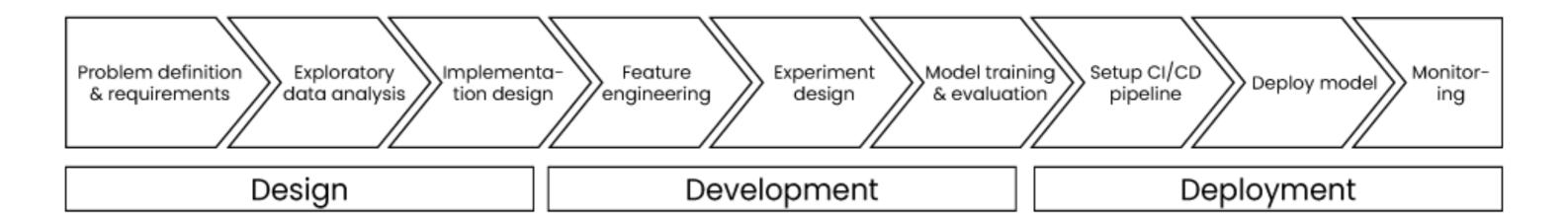
### Machine learning design



- · Added value
- Business requirements
- Key metrics
- Data processing

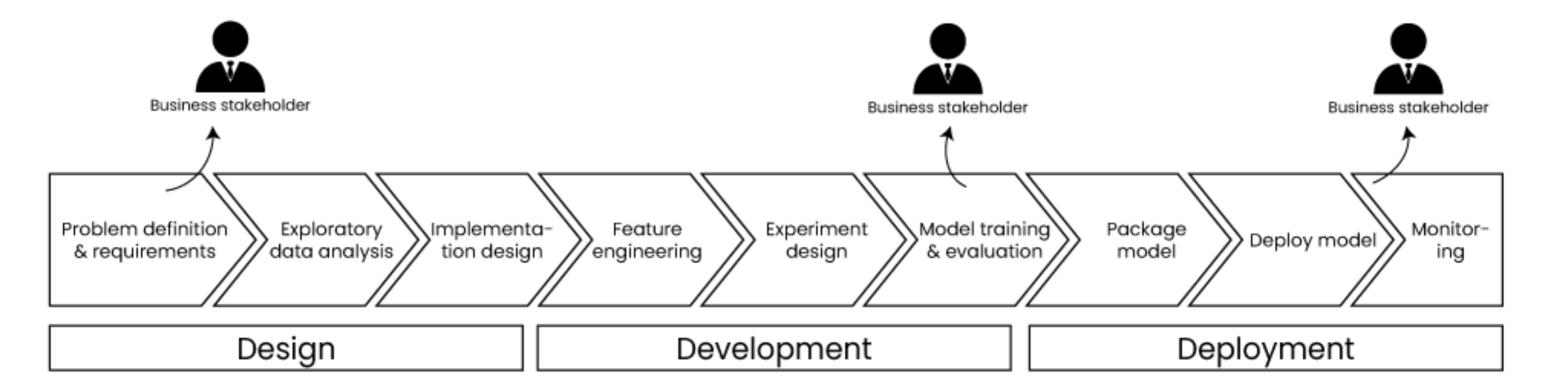
- Feature engineering
- Experiment tracking
- Model training & evaluation
- Runtime environments
- Microservices architecture
- CI/CD pipeline
- · Monitoring & retraining

## Machine learning lifecycle



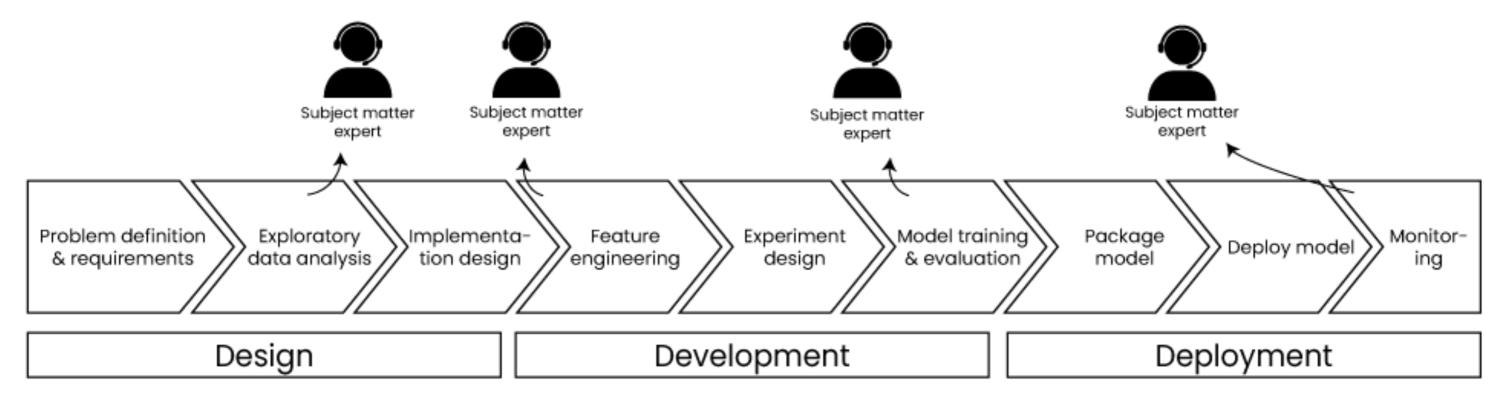
- Business roles
- Technical roles

### Business roles: business stakeholder



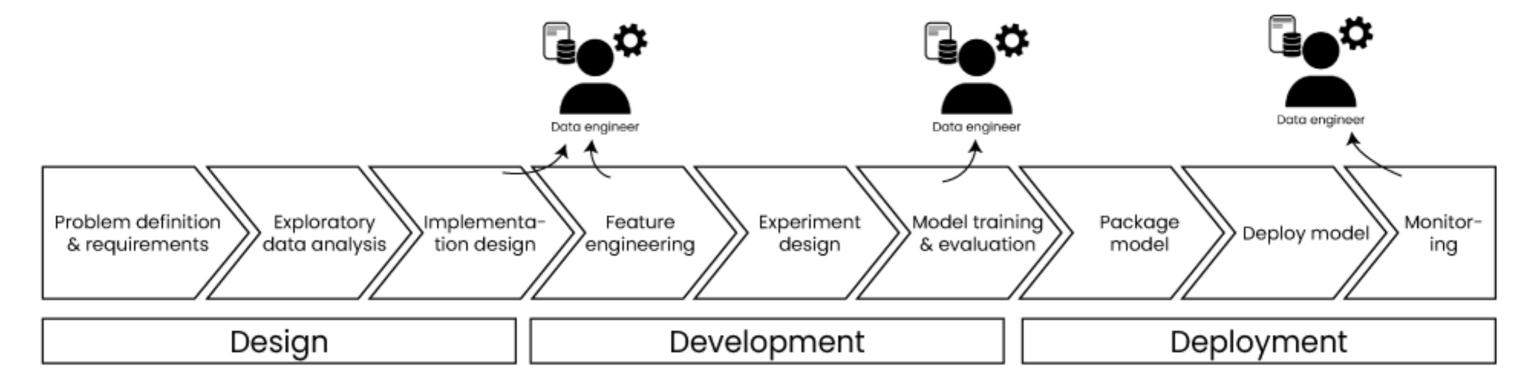
- Budget decisions
- Vision of company
- Involved throughout the lifecycle

### Business roles: subject matter expert



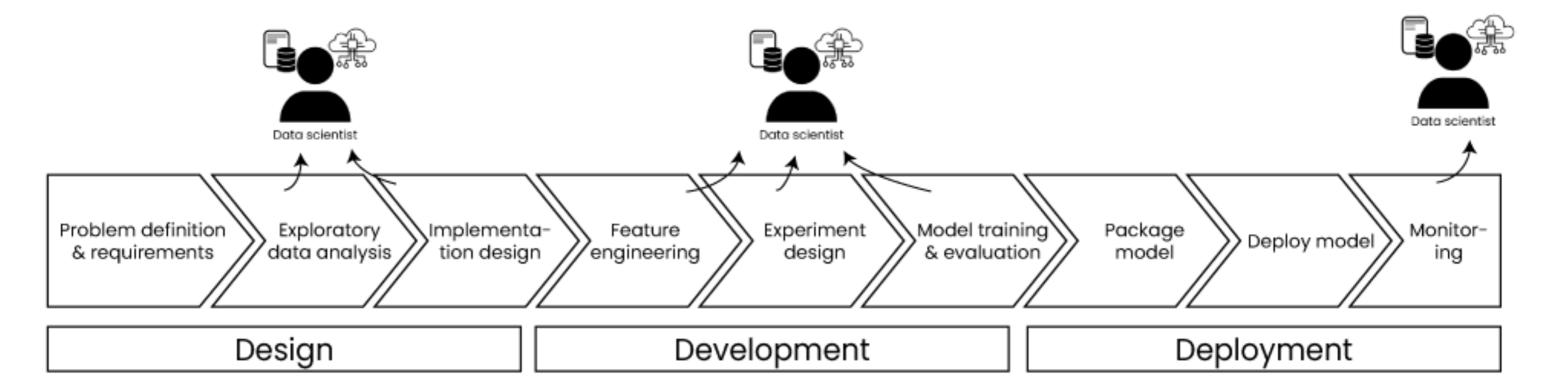
- Domain knowledge
- Involved throughout the lifecycle
- Interpret and validate data

### Technical roles: data engineer



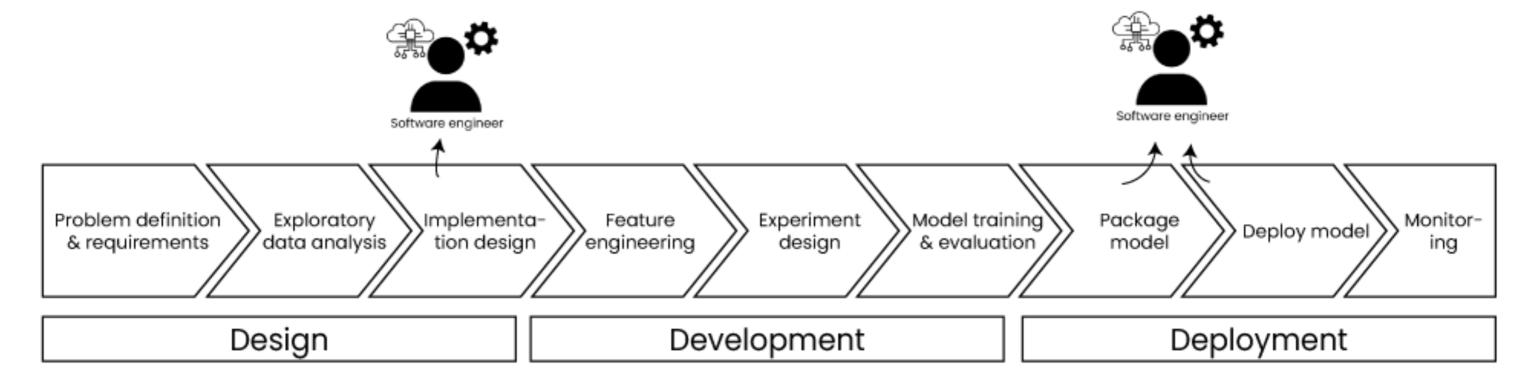
- Collecting, storing, and processing data
- Check and maintain data quality

### Technical roles: data scientist



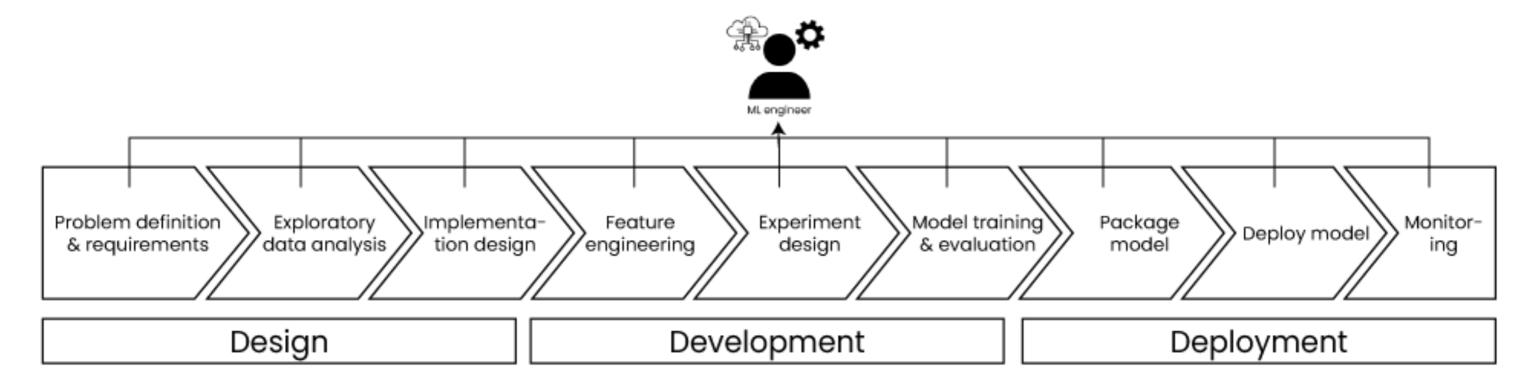
- Data analysis
- Model training and evaluation

### Technical roles: software engineer



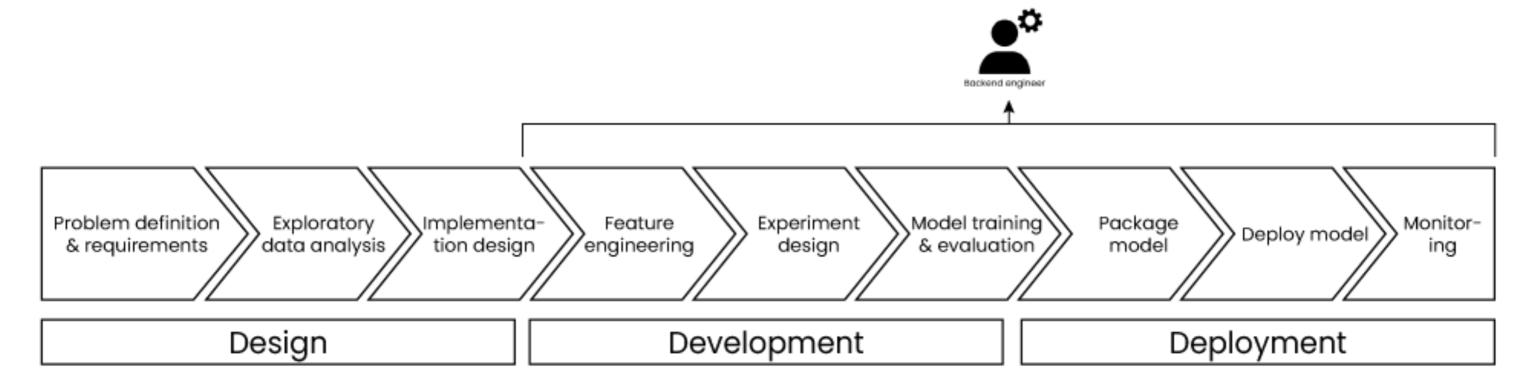
- Write software for model deployment
- Make sure that code follows guidelines

### Technical roles: ML engineer



- Versatile role
- Specifically designed for complete machine learning lifecycle

## Technical roles: backend engineer



- Enable development and deployment
- Cloud infrastructure

#### **Technical Debt**

the implied cost of additional rework caused by choosing an easy (limited) solution now instead of using a better approach that would take longer

~ Wikipedia[1]

#### Famous Google paper on the topic:

"Machine Learning: The high-interest credit card of technical debt" [2]

#### MORE TIME AND MODELS DEPLOYED

- More technical debt and model risk
- Process slower, more frustrating and error prone
- Ability to deliver value impaired

<sup>1</sup> https://en.wikipedia.org/wiki/Technical\_debt <sup>2</sup> https://research.google/pubs/pub43146/



#### ML workflows

- Data collection and preparation
- Data-labeling
- Model selection
- Model training
- Model packaging
- Model deployment
- Model monitoring and maintenance

ML workflow automation == MLOps maturity

